



Predicting Obesity using Machine Learning

Problem Statement

1 Identify Risk Factors

Uncover the key factors that contribute to obesity, such as diet, physical activity, demographics, and lifestyle habits.

2 Predict Obesity Prevalence

Develop a machine learning model that can accurately predict the likelihood of an individual becoming obese.

3 Inform Prevention Strategies

Leverage the insights gained to design targeted interventions and public health initiatives to combat the obesity epidemic.

Obesity rates

As % of total adult population (aged 15 years and over), 2015 or nearest year



* means that self-reported height and weight data are used in these countries, while measured data in other countries.
Source: OECD (2017), OECD Health Statistics 2017 (Forthcoming in June 2017).
oecd.org/health/obesity-update.htm



Dataset Overview

Data Source

The dataset is from a Kaggle competition, providing comprehensive information on individual health and lifestyle factors.

Features

The dataset includes variables such as age, gender, height, weight, physical activity, diet, and socioeconomic status.

Target Variable

The primary goal is to predict the binary outcome of whether an individual is obese or not.

Data Preprocessing

1

Data Cleaning

Identify and handle missing values, outliers, and inconsistencies to ensure data integrity.

2

Feature Scaling

Apply appropriate scaling techniques, such as standardization or normalization, to ensure all features are on a similar scale.

3

Encoding Categorical Variables

Convert categorical variables into a format suitable for machine learning algorithms, such as one-hot encoding.



Feature Engineering

Derived Features

Create new features by combining or transforming existing variables, such as BMI or waist-to-hip ratio.

Dimensionality Reduction

Apply techniques like PCA or feature selection to identify the most informative features and reduce complexity.

Domain Knowledge

Leverage subject matter expertise to engineer features that capture important biological, behavioral, and environmental factors.

Feature Importance

Analyze the relative importance of each feature to understand the key drivers of obesity and guide model interpretation.

Model Selection and Training



Classification Models

Explore a variety of supervised learning algorithms, such as logistic regression, decision trees, and random forests.



Ensemble Methods

Analyzing obesity dataset with individual factors using machine learning. Aim is trained model ense



Model Validation

Employ techniques like cross-validation to ensure the models generalize well to unseen data.

Model Evaluation

Accuracy	Measures the overall correctness of the model's predictions.
Precision	Evaluates the model's ability to correctly identify positive instances.
Recall	Assesses the model's ability to detect all positive instances.
F1-Score	Combines precision and recall to provide a balanced performance metric.



Insights and Recommendations

1

Key Risk Factors

Identify the most significant contributors to obesity, such as sedentary lifestyle, unhealthy diet, and socioeconomic status.

2

Targeted Interventions

Develop tailored prevention and treatment strategies based on the identified risk factors.

3

Policy Implications

Inform public health policies and initiatives to address the systemic drivers of the obesity epidemic.